Computation with the KEGG pathway database

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Abstract

We introduce and discuss a new computational approach towards prediction and inference of biological functions from genomic sequences by making use of the pathway data in KEGG. Due to its piecewise nature, the current approach of predicting each gene function based on sequence similarity searches often fails to reconstruct cellular functions with all necessary components. The pathway diagram in KEGG, which may be considered a wiring diagram of molecules in biological systems, can be utilised as a reference for functional reconstruction. KEGG also contains binary relations that represent molecular interactions and relations and that can be utilised for computing and comparing pathways. © 1998 Elsevier Science Ireland Ltd. All rights reserved.

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1. Introduction

Living organisms behave as complex systems which are flexible and adaptive to their surroundings. At the molecular level, organisms consist of intricate networks of molecular reactions, which are often called biochemical pathways. Scientists in the fields of biochemistry, molecular genetics, and molecular biology have made great endeavours to uncover various types of biochemical pathways and to abstract common features in different pathways and in different organisms. The best characterised among them are the metabolic pathways that involve enzymatic reac-

tions of chemical compounds. While computational biologists have long been tackling the problem of modelling, simulating, and predicting the metabolic pathways (Franco and Canela, 1984; Brutlag et al., 1991; Cohen and Bergman, 1994; Hofestädt and Meineke, 1995; Okamoto et al., 1996), the recent explosion of genomic sequence data poses new problems and raises new interests, which is the subject of this manuscript.

At the time of this writing, the complete genomic sequences of nine species are made publicly available. Genomic sequencing has become essential towards elucidating and understanding complicated molecular systems in a cell. Although systematic biochemical and genetic experiments will be necessary in the following phase of functional genomics, computational approaches are

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also expected to take a large and complementary part of the functional characterisation. During the last decade, the sequence databases, combined with the progress in sequence similarity search methods, have proved extremely useful for functional prediction of a single gene or a single molecule. However, in order to characterise the cellular function that results from a network of interacting molecules, a new 'pathway' database must be developed for comparison and computation at a higher level of the biological system. Basically, there are two major problems in representing and computing biochemical pathways.

First, the vast amount of knowledge on molecular pathways that has been accumulated for different cells and different organisms is underrepresented in databases and dispersed over literature. Even in limited attempts to computerise such knowledge, the pathway data are usually entered in databases primarily for browsing purposes. In contrast, we consider that the pathway knowledge should be computerised for the purposes of computation. It is expected that there will be complete genomes of hundreds of species from the three domains of life, *Archaea*, *Bacteria*, and *Eucarya* (Woese et al., 1990; Koonin, 1997). The in silico reconstruction of metabolic pathways is already an essential tool for functional assignment of predicted genes, for almost no data exist by biochemical experiments (Mushegian and Koonin, 1996; Danchin, 1997).

Second, new bioinformatics technologies need be developed to assist functional prediction from sequence data, especially by incorporating systems views on molecular pathways and molecular assemblies. It should be emphasised that the complete genomes of yeast and several bacteria still contain one third to over one half of ORFs that are left uncharacterised because no significant hits are found with well characterised sequences in the existing databases. The threshold of significant hits is somewhat arbitrary and it is often determined with consideration on the expected ratio of true predictions to false predictions (Brenner et al., 1995; Hubbard, 1996). However, a more desirable bioinformatics approach should mimic and automate the biologists' reasoning steps to effectively decrease the threshold when additional bio-

logical information is found to be associated with the sequence similarity.

KEGG (Kyoto Encyclopedia of Genes and Genomes) is our new bioinformatics project initiated in 1995 at Kyoto University (Kanehisa, 1997a). KEGG aims at:

- organising and computerising all current knowledge of molecular and genetic pathways from experimental observations,
- maintaining the gene catalogue of every organism that has been sequenced and mapping each gene product onto a component in the pathway, and
- developing new bioinformatics technologies for comparing and computing pathways.

The thrust of the project is to describe, utilise, predict, and possibly design systems behaviours of living organisms. Here we report the computational methods that have been developed to efficiently utilise the metabolic and genomic data in KEGG.

2. Pathway database

The KEGG pathway database consists of two sections: the metabolic pathway section and the regulatory pathway section. While the current knowledge of metabolic pathways is well organised in KEGG, the organisation of regulatory pathways is still rudimentary. The metabolic pathway data were first entered from the book 'Metabolic Maps' compiled by the Japanese Biochemical Society (Nishizuka, 1980, 1997) and the wall chart of 'Biochemical Pathways' by Boehringer-Mannheim (Gerhard, 1992). They were then verified and updated by using mostly the EMP/WIT database (Selkov et al., 1996) and the 13 volume Enzyme Handbook (Schomburg and Salzmann, 1990).

At the moment KEGG contains about 100 metabolic pathway diagrams grouped into ten categories. Each diagram can be retrieved by the hierarchical text menu, by the hierarchically drawn graphics menu, or by the key word search using DBGET/LinkDB (Kanehisa, 1997b) via the KEGG WWW server (http://www.genome.ad.jp/kegg/). Fig. 1 shows an example for cysteine
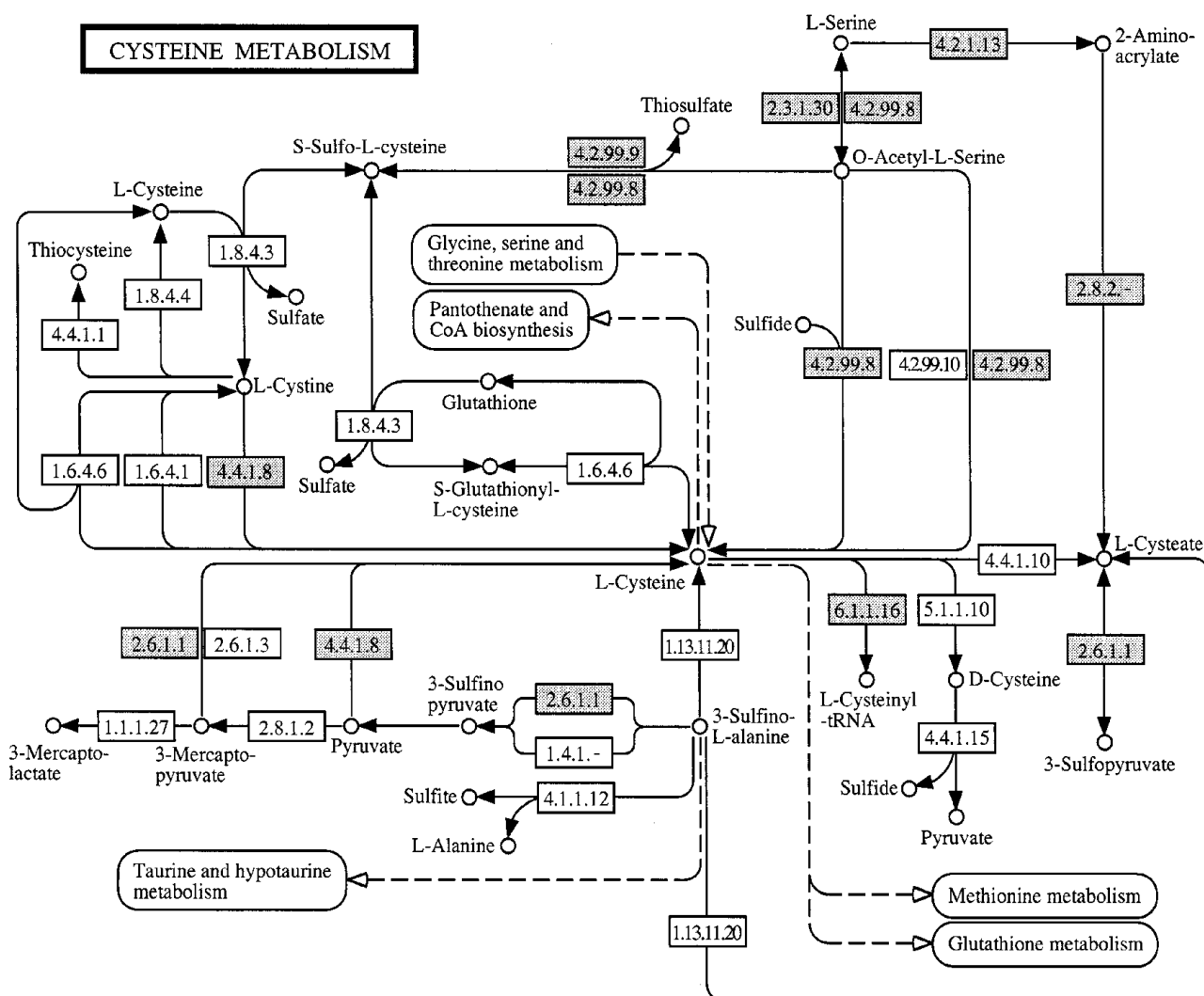
Fig. 1. An example of the KEGG pathway diagram that represents the relationships of genes or gene products, in this case, in the pathway of cysteine metabolism. A rectangle is a gene product (an enzyme) and it is marked (shaded) when the corresponding gene is found in the genome, in this case, of *E. coli*.

metabolism. There are two characteristics of the KEGG metabolic pathway diagrams that are relevant to computation.

First, an enzyme denoted by a rectangle is an object that can be manipulated by machine. For each pathway diagram there is one reference diagram which is manually drawn and updated, and all organism specific diagrams are computationally derived by matching the enzyme objects and the corresponding genes in the gene catalogue. In this way KEGG is able to immediately reconstruct organism-specific pathways for an increasing number of complete genomes. The enzyme

object is hyperlinked to the enzyme section of the ligand database containing, among others, the enzyme nomenclature, the reaction scheme, the chemical compounds involved, and additional links to molecular and biological information (Suyama et al., 1993; Goto et al., 1998).

Second, the pathway diagram is drawn to lay stresses both on the successive conversion of major compounds and on the appearance of two consecutive enzymes in the pathway. The information is computerised in the form of, what we call, binary relation, which is the most fundamental data type in KEGG. The substrate-product
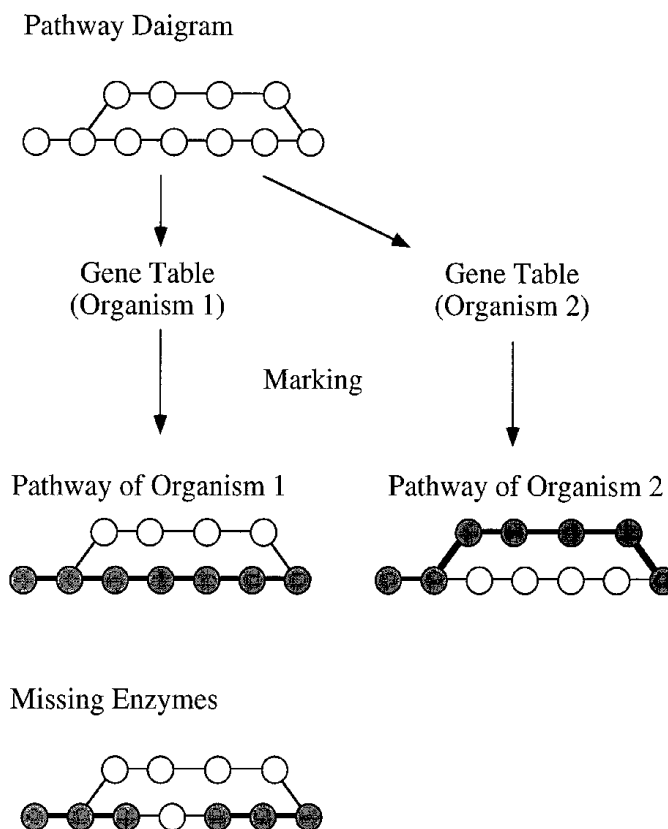
Pathway Daigram

Fig. 2. A schematic illustration of generating organism-specific pathways. The reference pathway diagram is marked by mapping the information in the gene table on to the components in the pathway, resulting in a sequence of marked genes that can be considered an organism-specific pathway. Missing enzymes are the unmarked components that split the organism specific pathway.

binary relations have been utilised in computing possible reaction paths (Goto et al., 1996) and the enzyme–enzyme binary relations have been correlated with other types of molecular relations (Ogata et al., 1996). A similar concept is the predecessor-list in EcoCyc (Karp et al., 1996) which is utilised for automatic drawing of metabolic pathways.

3. Pathway computation

3.1. Reconstructing organism-specific pathways

An organism-specific pathway is automatically generated by matching the reference pathway diagram and the gene catalogue according to the EC number. When the gene for an enzyme exists in the gene catalogue, the box representing the corresponding enzyme is marked by colour on the

pathway. The consecutive appearance of the coloured boxes would then be considered an organism-specific pathway (Fig. 2). In order for this procedure to be successful, the reference diagram should contain all known alternatives of reaction paths rather than just the consensus alone. In fact, we are learning from the complete genomes that different organisms have different sets of metabolic pathways that reflect the living environment and the strategy of adaptation. The automatic reconstruction of organism-specific pathways is a first step toward establishing a well verified set of reference diagrams.

It is often the case that, when all the predicted enzyme genes are mapped on the KEGG reference pathway diagram, the organism-specific pathway would not become continuous or complete because of missing enzymes (Fig. 2). It is possible that the initial gene finding and functional assignment have not been correctly done

and missed the enzyme that is actually present in the genome. By changing the threshold values and other parameters the gene may be identified. Alternatively, our knowledge on enzymatic reactions is not sufficient, and an enzyme with a given EC number may function for substrates with wider specificity than previously known. In fact, the number of aminotransferases that appear at the last step of amino acid biosynthesis seemed different in different bacteria and it was necessary to assume that, for example, aspartate aminotransferase catalyse reactions for other amino acids as well (Bono et al., 1998). Moreover, there is also a possibility that our knowledge on metabolic pathways is not sufficient, and alternative paths of reactions may lead to the conversion of the initial substrate to the final product in the missing portion. We formulate this problem in terms of logic programming as described below.

3.2. Computing paths of chemical reactions

In general an enzymatic reaction involves multiple substrates and multiple products. In KEGG a reaction is represented by a collection of binary relations between all possible substrate–product pairs (Goto et al., 1996). Here we consider the simplest case of one substrate—one product reaction. As formalised below, the reaction catalysed by a specific enzyme is a factual data, while the path is defined in the form of a rule:

$reaction\ (X,\ Y,\ E)$
$path\ (X,\ Y,\ [E]) \leftarrow reaction(X,\ Y,\ E)$
$path\ (X,\ Y,\ [E\ |\ EL]) \leftarrow reaction(X,\ Z,\ E),$
$$path(Z,\ Y,\ EL)$$

where $X,\ Y,\ Z$ are the chemical compounds, E is the enzyme, and EL is the list of enzymes. The rule says that the path can be composed of just one reaction or it can be derived by extending an existing path by one reaction. As will become apparent in Section 3.3 this formalisation is not limited to chemical reactions; namely, successive additions of binary relations can be used to deduce different types of paths.

We have developed a path computation program, PathComp, to efficiently perform linking of binary relations (Goto et al., 1996; Fujibuchi et

al., 1997). The implemented algorithm is based on a breadth first search and PathComp provides several options for the calculation. For example, it calculates all paths starting from a given initial substrate, all paths between an initial substrate and a final product, and the shortest paths along a set of compounds. PathComp generates candidates for organism-specific pathways or all chemically feasible pathways either by limiting the reaction data for the enzymes in the gene catalogue or by using all the reaction data extracted from the LIGAND database, respectively.

As mentioned, the substrate specificity of an enzyme in one organism could be different from that of the same enzyme in another organism. According to the hierarchical classification of EC numbers, this implies that it may be necessary to include other reactions in the same category for the enzyme gene predicted from the genome. This is defined by another rule:

$reaction(X,\ Y,\ E') \leftarrow reaction(X,\ Y,\ E),$
$$group(G,\ E,\ E')$$

where enzymes E and E' belong to the same group G, and this procedure is called query relaxation (Gaasterland et al., 1992). In PathComp the relaxation can be done according to the EC number classification, the grouping of sequence motifs, the superfamily (sequence similarity) classification and the 3D-fold classification.

3.3. Computing paths of reasoning

Various kinds of reasoning steps during in silico analysis taken by biologists often involve sophisticated navigation across many links among biological entities. The nature of these links would be classified into three categories shown in Table 1: factual, similarity, and biological links (Kanehisa, 1997b). Automation of the reasoning steps by path computation of these links is one of the major objectives of the KEGG project. The synthesis of metabolic pathways described above is a straightforward application of utilising a well-defined category of biological links, namely, the substrate-product binary relations.

The similarity search against sequence databases is currently the only automated portion to

Table 1
Three type of links used in path computation

Type	Example
Factual link	Cross-references in databases
	Links between genes and functions
Similarity link	Sequence homology (orthology/paralogy)
	3D similarity and complementarity
Biological link	Substrate–product relations in enzymatic
	reactions
	Interacting molecules in a cell
	Neighbouring genes in the genome

assist biological reasoning for newly determined sequences. Although the links to other databases may be provided, the biological information associated with the similarity found is strongly dependent on texts that can only be understood by human. KEGG attempts to automate the latter portion of reasoning by computerising different types of biological links, including molecular counterparts in the pathway or closely associated genes in the genome. Although sequence similarity is just another type of binary relation in KEGG, the entire paths of reasoning can be computed in a similar way as the path computation of chemical reactions.

Fig. 3 shows a specific example, where the reasoning step is defined as a sequence of combining three different links. Suppose that any functional clue is sought for a newly determined human gene sequence. By using the similarity links of this gene, homologues in other organisms can be found. The biological links provided by KEGG would then identify biological partners in the pathways of these organisms. By using the
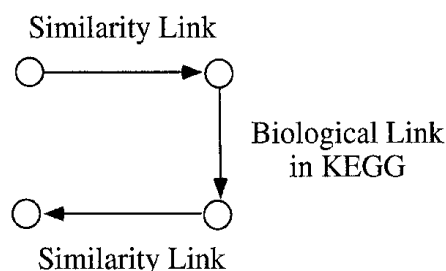

Fig. 3. An example of the paths of reasoning that can be computed with a combination of various kinds of links.
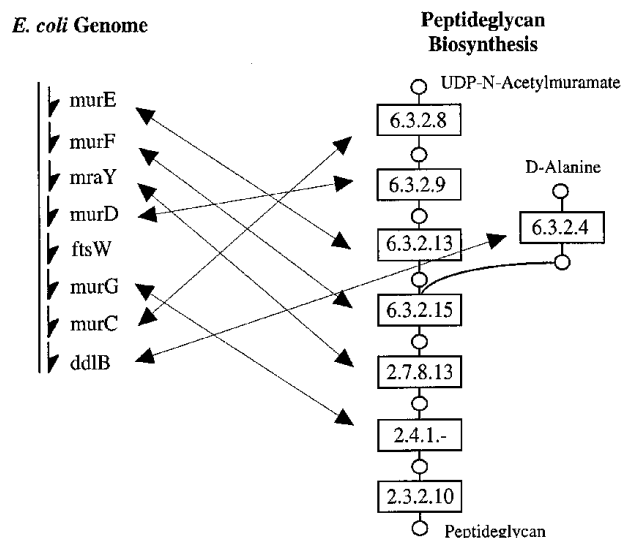

Fig. 4. The cluster of genes in the *E. coli* genome is often well correlated with the cluster of enzymes in the metabolic pathway as in this case of peptidoglycan biosynthesis. This type of correlation can be examined graphically in KEGG.

similarity links back to human, it is possible to find candidate genes or gene products that may be interacting with the initial gene or the gene product in the biological pathways. This and other types of reasoning steps are currently being implemented by using the DBGET/LinkDB system (Fujibuchi et al., 1997). However, the most critical aspect is the organisation and computerisation of actual data on biological links that are dispersed over organisms and over literature. The current version of DBGET/LinkDB (http://www.genome.ad.jp/dbget/dbget.links.html) contains only the enzyme-enzyme relations in the metabolic pathways.

Chromosomal locations of genes is also informative for elucidating regulation or evolution of metabolism in yeast and bacteria (Wolf and Butler, 1997; Otsuka et al., 1996). In KEGG the location and the order of enzyme genes in the genome can be examined graphically, as well as the location and the order of enzymes in the pathway. The example in Fig. 4 shows that a set of enzymes coded in close positions along the *Escherichia coli* genome forms a block of reaction paths in the metabolic pathway, which is actually a general tendency as described below. Although the feature is not provided yet, the positional

correlation in the genome, which is also represented by binary relations, will be included in the automatic biological reasoning process in KEGG.

3.4. Comparing biological networks

The previous two Sections 3.2 and 3.3 described apparently the same kind of computation. Here we present another type of computation, which can be categorised as the comparison of biological networks. Here the network includes not only the metabolic and regulatory pathways, but also the genome which is a one-dimensional network of genes and the hierarchical network of evolution formed by orthologous genes. By comparing the same type of networks, for example, the metabolic pathway diagrams, it is possible to identify similarities and variations among different species. By comparing different types of networks, for example, the metabolic pathway and the genome, it is possible to identify groups or clusters of genes that may be considered modules of the biological system.

In bacterial genomes it is known that functionally related genes tend to be clustered more often than unrelated genes (Tamames et al., 1997). We searched the enzymes that are coded in close positions along the genome and that play their roles in close positions on the metabolic pathways. The functionally related enzymes coding segments (FRECSs), as we call them, are obtained as shown in Fig. 5(a). The path length between two enzymes is defined by the shortest path calculation of the PathComp program using the enzyme-enzyme binary relations that represent the knowledge of actual metabolic pathways in KEGG. The path length is calculated for each enzyme pair that appears within the sliding window along the genomic sequence. Fig. 5(b) shows the mean path length of the enzyme pairs within the window sizes of 1 and 10 kbp for the *E. coli* genome that contain, on the average, one gene per 1 kbp. First of all, the average value for all windows decreased from 5.7 (dotted line) to 4.5 (solid line) with the decrease in the window size. We considered that this correlation would be due to the existence of relatively short (~ 10 kbp) clusters of related enzyme genes. By marking the related enzyme

gene pairs with the path length of three or less and by merging the overlapping windows of 10 kbp containing related enzyme genes, we obtained the total of 82 FRECSs. Among them about 54% were found to be related to known operons. The enzymes in the peptideglycan biosynthesis shown in Fig. 4 was one of the examples.
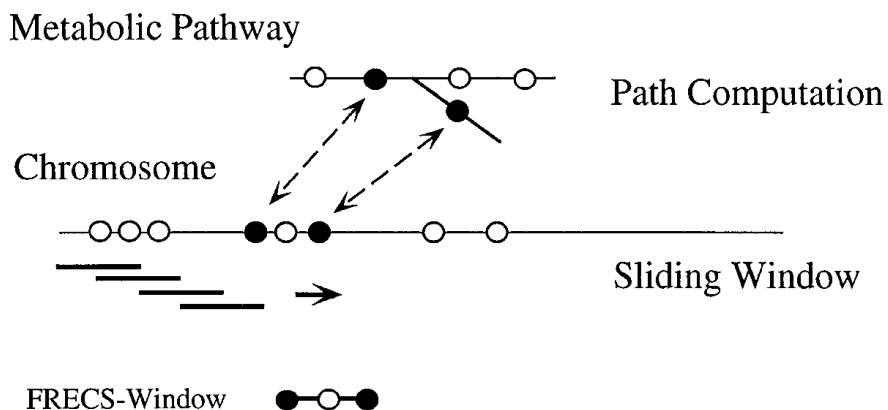
Although this analysis focuses on a comparison of clustering based on two biological links representing pathways and genomes, a similar comparison can be applied to other types of links and networks. We observed the correlation of clusters between the biological links of metabolism and the sequence similarity links; namely, there was a tendency of paralogous genes appearing closely in the metabolic pathway (Ogata et al., 1996). There were also cases where apparently similar sets of enzymes formed similar reaction paths when local similarity patterns of metabolic pathways were searched. We are extending this type of network comparison for analysis of other problems involving evolution (sequence similarity), pathway formation, and genome organisation.
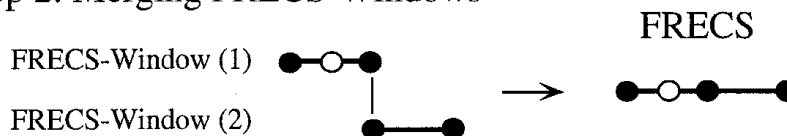
4. Discussion

In this paper, we have described a computational approach that is intended to assist in silico investigation of gene functions and molecular pathways. This has become feasible because of the unique pathway database being organised by KEGG. A biochemical pathway is an abstraction of a subset of intricate networks in the soup of interacting biomolecules (Mavrovouniotis, 1995), and the abstraction is arbitrarily chosen by biologists at the level of their interests. KEGG's pathway is at a highest level of abstraction where only the relation of gene products is emphasised without much describing the detailed molecular processes involved. This is because KEGG's major concern is to cover all aspects of biochemical pathways and to correlate them with the genomic information. This breadth first approach is suitable for analysing overall behaviours of a biological system, and is complementary to the traditional depth first approach for analysing specific components of a biological system.

(a)

Step 1: Identification of FRECS-Windows

Metabolic Pathway

Path Computation

Chromosome

Sliding Window

FRECS-Window

Step 2: Merging FRECS-Windows

FRECS
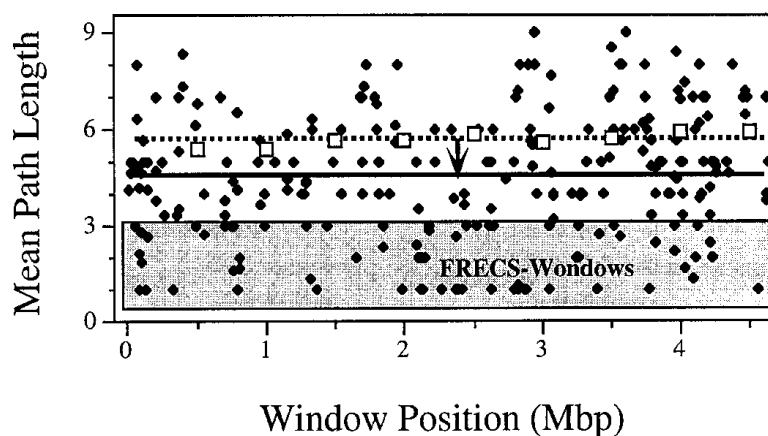
FRECS-Window (1)

FRECS-Window (2)

(b)

Fig. 5. (a) A functionally related enzymes coding segment (FRECS) is defined by the enzyme pairs that appear within a sliding window in the genome and that are separated by given path lengths in the metabolic pathway (see text). (b) The mean path length of enzyme pairs is plotted against the window position for two different window sizes: 1 Mbp (white boxes) and 10 kbp (black boxes). The dotted and solid lines show the average levels of the path length for 1 Mbp and 10 kbp windows, respectively, and an arrow is put to indicate the difference of the levels. FRECS corresponds, in this case, to the path length of 3 or less.

Somewhat similar to the level of abstraction, the experimental data should also be viewed in a hierarchical way. In contrast to the traditional experiments in molecular biology, the experiments such as the systematic analysis by the yeast two-hybrid system (Luban and Goff, 1995) produce, more or less, qualitative data that are accumulated in much larger amounts and at much faster rates. In order to capture an overall picture of the biological system, the progress in data representation and computation of such kinds of data is necessary.

Most of the current computational tools in molecular biology are strongly dependent on the structural information of molecular sequences and 3D co-ordinate data that are relatively simple to represent. In contrast, it is not an easy task to represent functional information for computation by machine, although it is not difficult to describe function by texts for human to understand. We consider the essence of function lies in the interaction of molecules and adopted the representation of binary relations. The network is an abstraction of molecular pathways, gene organisation in the genome, and functional and structural hierarchies of molecules, all of which is considered to result from a set of binary relations. Furthermore, the network is also an abstraction of the chain of logical reasoning steps by human, or the deduction from a set of relations. Thus, the path computation techniques and the network comparison techniques based on binary relations will have a broad range of applications.

The benefit of whole genome sequencing is the completeness of the ORF catalogue, even though it may contain uncharacterised ORFs and may suffer from subsequent additions, deletions, and modifications. In general, incompleteness of a cellular subsystem defined by a set of genes or gene products taken from the complete catalogue is quite suggestive for further investigation, as in those missing enzymes in a specific metabolic pathway. The concept of the minimal gene set is another example of the power of cross-species comparison of complete catalogues (Mushegian and Koonin, 1996). As the computational analyses play major roles in functional genomics, the management of the functional, albeit predicted,

data also requires a major investment because of frequent updates. In addition to the PATHWAY database, KEGG maintains the GENES database that is a collection of gene catalogues for many organisms. Each gene catalogue is hierarchically classified according to the most up-to-date functional assignment and all the catalogues are correlated by the orthologous gene table. With the development of the KEGG databases, the computational techniques such as for path computation and network comparison will become increasingly useful in automating reasoning steps that mimic biologists' logics in silico.

Acknowledgements

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